Documentation: Surya's Text to Math Problem Solver and Data Research Assistant Using Streamlit

Overview

This project is a **Streamlit-based chatbot** that assists users in solving mathematical problems step-by-step and retrieving relevant data from Wikipedia. The chatbot uses **Groq's AI models** to process user queries and provides detailed explanations for mathematical solutions.

Features

1. Text-to-Math Problem Solving

- Accepts a **math problem** from the user.
- Converts the problem into a **numerical expression** and solves it.
- Provides step-by-step explanations for the solution.

2. Wikipedia Data Retrieval

- Uses Wikipedia API to fetch relevant information.
- Helps with **concept explanations** related to math problems.

3. Intelligent Agent for Answering Queries

- Uses **LLM-based tools** to process questions.
- Can handle **complex problem-solving tasks** with logical reasoning.

Architecture & Components

1. Streamlit UI

• Sidebar for entering the Groq API Key.

- Main interface to enter a math question and view responses.
- Displays **chat history** with user and assistant messages.

2. Language Model (LLM) - ChatGroq

- The system connects to **Groq's AI models** for generating responses.
- The model used is "gemma2-9b-it".

3. Wikipedia API Wrapper

- Enables the assistant to **search Wikipedia** for additional knowledge.
- Used as a tool to **supplement answers** when needed.

4. Math Processing System

- Uses an **LLM-based Math Chain** to evaluate **numerical expressions**.
- A custom **prompt template** is designed to ensure valid calculations.
- Extracts the **mathematical expression** from the user's question.
- Handles cases where **variables** need to be assigned values.

5. Reasoning System

- A dedicated tool for step-by-step logical reasoning.
- Breaks down the solution into a **structured explanation**.
- Uses a **custom prompt** for structured reasoning responses.

6. Al Agent Integration

- Combines multiple tools (Wikipedia, Math Solver, Reasoning) into a single intelligent assistant.
- Uses Zero-Shot ReAct Agent for decision-making.
- Handles parsing errors gracefully.

How It Works

- 1. The user enters a **math problem** in the text box.
- 2. The chatbot determines whether to:
 - Calculate directly (if it's a straightforward math question).
 - Use logical reasoning (if explanation is needed).

- **Fetch data from Wikipedia** (if additional knowledge is required).
- 3. The response is displayed with:
 - Numerical expression extracted from the question.
 - Step-by-step explanation of the solution.
 - **Final result** with detailed insights.

Error Handling & Robustness

- Validates API Key before running the chatbot.
- **Ignores invalid mathematical expressions** to prevent errors.
- Handles missing numerical values by making reasonable assumptions.
- Filters Wikipedia results to return only relevant information.
- **Displays error messages** if something goes wrong.

User Interaction Flow

- 1. User enters a math problem.
- 2. The assistant analyzes the question.
- 3. If it's a direct calculation, the math solver is used.
- 4. If **reasoning is required**, the **reasoning tool** generates a structured explanation.
- 5. If additional information is needed, the Wikipedia tool fetches related concepts.
- 6. The response is displayed in a **chat-style format**.

Possible Use Cases

- Students learning math who need step-by-step explanations.
- Teachers & tutors who want to verify and explain math solutions.
- **Researchers** looking for quick Wikipedia-based references.
- Casual users who need help solving math problems.

Limitations

- Requires a valid Groq API Key to function.
- Assumptions for **unknown variables** may not always match user expectations.
- Wikipedia tool is **limited to English-language articles**.
- May not handle highly advanced symbolic algebra.

Future Enhancements

- Integrating a **graphing feature** for visual explanations.
- Supporting LaTeX rendering for math expressions.
- Enhancing Wikipedia search with multi-source document retrieval.
- Adding **voice input** for a more interactive experience.

Conclusion

This **AI-powered assistant** simplifies complex math problems and provides **detailed**, **step-by-step solutions**. It integrates **Wikipedia search** and **logical reasoning** to enhance learning and research. Built with **Streamlit**, this tool makes math problem-solving **intuitive and user-friendly**.

Code:

import streamlit as st

from langchain_groq import ChatGroq

from langchain.chains import LLMMathChain, LLMChain

from langchain.prompts import PromptTemplate

from langchain_community.utilities import WikipediaAPIWrapper

from langchain.agents.agent_types import AgentType

from langchain.agents import Tool, initialize_agent

```
from doteny import load_doteny
from langchain.callbacks import StreamlitCallbackHandler
import re
#Load environment variables
load_dotenv()
# Setting up the Streamlit app
st.set_page_config(page_title="Surya's Text to Math Problem Solver and Data Research Assistant")
st.title("Text to Math Problem Solver")
#Get Groq API Key from sidebar
groq_api_key = st.sidebar.text_input(label="Groq API Key", type="password")
if not groq_api_key:
 st.info("Please type your GROQ API key to continue")
 st.stop()
# Initialize the language model
try:
 llm = ChatGroq(model="gemma2-9b-it", groq_api_key=groq_api_key)
except Exception as e:
 st.error(f"Failed to initialize model: {e}")
 st.stop()
# Initialize the Chat Tools
```

```
wikipedia_wrapper = WikipediaAPIWrapper()
wikipedia_tool = Tool(
 name="Wikipedia",
 func=wikipedia_wrapper.run,
 description="A tool for searching Wikipedia to assist with math problems"
# Initialize the Math tool with enhanced explanation
try:
 math_chain = LLMMathChain.from_llm(llm=llm)
 # Define a custom math prompt to enforce numerical expressions
 math_prompt = """
 You are a mathematical assistant. For the given question, provide a numerical expression (no
variables) and a detailed, point-wise explanation of how to solve it. If the question contains
variables, assume reasonable numerical values (e.g., k=1) and state your assumption:
 Question: {question}
 Numerical Expression: <expression>
 Description:
 - Step 1: [First step]
 - Step 2: [Second step]
 - ... [Continue as needed]
```

```
Final Result: <result>
 math_prompt_template = PromptTemplate(input_variables=["question"], template=math_prompt)
 math explain chain = LLMChain(llm=llm, prompt=math prompt template)
 def calculate_with_explanation(question):
    try:
      # Get explanation from LLM
      explanation = math_explain_chain.run({"question": question})
      #Extract numerical expression
      expr_match = re.search(r'Numerical Expression: (.*?)\n', explanation)
      if not expr_match:
         return f"Question: {question}\nError: No valid numerical expression
provided\nDescription:\n- Step 1: Failed to parse a numerical expression\nFinal Result: N/A"
      expr = expr_match.group(1).strip()
      #Check for variables and handle them
      if re.search(r'[a-zA-Z]', expr):
        assumption = "Assumption: Any variables (e.g., k) set to 1 unless specified."
        expr = re.sub(r'[a-zA-Z]', '1', expr) # Replace variables with 1
        explanation = f'' \{explanation\} \setminus n \{assumption\}''
      result = math_chain.run(expr)
```

```
return f"{explanation}\nFinal Result: {result}"
    except Exception as e:
      return f"Question: {question}\nNumerical Expression: {expr if 'expr' in locals() else
'N/A' \nDescription:\n- Error: {e \\nFinal Result: N/A"
  calculator = Tool(
    name="Calculator",
    func=calculate_with_explanation,
    description="Solves math questions with step-by-step explanations."
  )
except Exception as e:
 st.error(f"Math tool initialization failed: {e}")
  st.stop()
# Define the reasoning prompt with step-by-step description
prompt = """
You are an agent tasked with solving the user's mathematical question. Provide a numerical
expression (no variables) and a detailed, point-wise explanation. If variables are present, assume
reasonable values (e.g., k=1) and note the assumption:
Question: {question}
Numerical Expression: <expression>
Description:
 Step 1: [First step]
```

```
Step 2: [Second step]
 ... [Continue as needed]
Final Result: <result>
prompt_template = PromptTemplate(input_variables=["question"], template=prompt)
#Combine tools into a chain
try:
 chain = LLMChain(llm=llm, prompt=prompt_template)
 reasoning_tool = Tool(
    name="Reasoning",
    func=chain.run,
    description="Answers math questions with step-by-step explanations."
 )
except Exception as e:
 st.error(f"Reasoning tool setup failed: {e}")
 st.stop()
# Initialize the agent
try:
 assistant_agent = initialize_agent(
    tools=[wikipedia_tool, calculator, reasoning_tool],
```

```
llm=llm,
    agent=AgentType.ZERO_SHOT_REACT_DESCRIPTION,
    verbose=False,
    handle_parsing_errors=True,
except Exception as e:
 st.error(f"Agent initialization failed: {e}")
 st.stop()
# Initialize session state for messages
if "messages" not in st.session_state:
 st.session_state["messages"] = [
    {"role": "assistant", "content": "Hi, I'm a Math Chatbot who can solve your math questions with
step-by-step explanations!"}
 ]
# Display chat history
for msg in st.session_state.messages:
 st.chat_message(msg["role"]).write(msg["content"])
# Function to generate the response
def generate_response(user_question):
 try:
```

```
response = assistant_agent.invoke({'input': user_question})
    resp = response['output'] if isinstance(response, dict) and 'output' in response else str(response)
    if "Description:" not in resp:
      return calculate_with_explanation(user_question) #Fallback for steps
    return resp
  except Exception as e:
    return f"Error: {e}"
# Interaction logic
question = st.text_area("Enter your Question:")
if st.button("Find my answer"):
 if question and question.strip():
    with st.spinner("Generating Response..."):
      st.session_state.messages.append({"role": "user", "content": question})
      st.chat_message("user").write(question)
      st_cb = StreamlitCallbackHandler(st.container(), expand_new_thoughts=False)
      response = generate_response(question)
      st.session_state.messages.append({"role": "assistant", "content": response})
      st.write("### Response:")
      st.success(response)
  else:
```

st.warning("Please enter a question")